

Patrolling a Path Connecting a Set of Points with Unbalanced Frequencies of Visits^{*}

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Abstract. Patrolling consists of scheduling perpetual movements of a collection of mobile robots, so that each point of the environment is regularly revisited by any robot in the collection. In previous research, it was assumed that all points of the environment needed to be revisited with the same minimal frequency.

In this paper we study efficient patrolling protocols for points located on a path, where each point may have a different constraint on frequency of visits. The problem of visiting such divergent points was recently posed by Gąsieniec et al. in [13], where the authors study protocols using a single robot patrolling a set of n points located in nodes of a complete graph and in Euclidean spaces.

The focus in this paper is on patrolling with two robots. We adopt a scenario in which all points to be patrolled are located on a line. We provide several approximation algorithms concluding with the best currently known $\sqrt{3}$ -approximation.

1 Introduction

In this paper we study efficient patrolling protocols by two robots for a collection of n points distributed arbitrarily on a path or a segment of length 1. Each

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point needs to be attended perpetually with *known* but often distinct minimal frequency, i.e., some points need to be visited more often than others.

The problem was recently studied in [13] where a collection of n points was monitored with use of a single mobile robot. The points to be patrolled in [13] are located in nodes of a complete graph with edges of uniform (unit) length, as well as in Euclidean spaces, where the points are distributed arbitrarily. In their work the frequency constraints refer to *urgency factors* h_i , meaning that during a unit of time the urgency of point p_i grows by an additive term h_i , and the task is to design a schedule of perpetual visits to nodes which minimizes the maximum ever observed urgency on all points. In complete graphs and for any distribution of frequencies (urgency factors) the authors of [13] proposed a 2-approximation algorithm based on a reduction to the pinwheel scheduling problem, see, e.g., [6,7,14,15,18]. They also discuss more tight approximations for the cases with more balanced urgency factors. In Euclidean spaces [13] proposes several lower bounds and concludes with an $O(\log n)$ -approximation for an arbitrary distribution of points and urgency factors.

In our formulation, we assume that both robots have unit speed, and we try to minimize the relative violation of visitation-frequency requirements, i.e. the worst case time between two visitations over the required largest waiting time of each point. Equivalently, one may think of the problem of finding the minimum possible speed s that both robots should patrol with that induces no violation for the visitation-frequency requirements. In such setting, our *patrolling* result refers naturally to a competitive ratio, which is defined by the ratio of the speed the robots attain in our algorithm divided by the speed in the optimal solution.

Specific to our model is the use of two robots, for which, as we show, one can achieve $\sqrt{3}$ -approximation patrolling schedules. Notably, and maybe counter-intuitively, reducing the number of robots from two to one does not lead to constant approximation. An instructive example is when the central point has a very large visiting frequency (we can dedicate one robot to this point) comparing to the rest of the points on the line.

In the previous research on boundary and fence patrolling (cf. [10,11,16]) all points of the patrolled environment were supposed to be revisited with the same frequency. However, assigning different importance to distinct portions of the monitored boundary seems natural and observable in practice. A particular variation of this problem was studied in [9], where the authors focus on monitoring *vital* (possibly disconnected) parts of a linear environment, while the remaining *neutral* portions of the boundary need not be attended at all.

The problem of distinct attendance assigned to different portions of the environment, while of inherent combinatorial interest, is also observed in perpetual testing of virtual machines in cloud systems [1]. In such systems the frequency with which virtual machines are tested for undesirable symptoms may vary depending on the importance of dedicated cloud operational mechanisms.

The problem studied here is also a natural extension of several classical combinatorial and algorithmic problems referring to *monitoring* and *mobility*. This includes the *Art Gallery Problem* [19,20] and its dynamic variant called the k -

Watchmen Problem [23]. In a more recent work on *fence patrolling* [9,10,16] the efficiency of patrolling is measured by the *idleness* of the protocol, which is the time where a point remains unvisited (maximized over all time moments and all points of the environment). In [11] one can find a study on monitoring of linear environments by robots prone to faults. In [10,16] the authors assume robots have distinct maximum speeds which makes the design of patrolling protocols more complex, in which case the use of some robots becomes obsolete.

In a very recent work [17] Liang and Shen consider a line of points attributed with uniform urgency factors. For robots with uniform speeds, they give a polynomial-time optimal solution, and for robots with constant number of speeds they present a 2-approximation algorithm. For an arbitrary number of velocities they design a 4-approximation algorithm, which can be extended to a 2α -approximation algorithm family scheme, where integer $\alpha > 1$ is the tradeoff factor to balance the time complexity and approximation ratio.

2 Problem Statement & Definitions

An instance of the *Path Patrolling Problem of Points with Unbalanced Frequencies* (PUF) consists of points $S = \{y_i\}_{i=1,\dots,n}$ in the unit interval, where $0 = y_1 < y_2 < \dots < y_n = 1$. Each point y_i is associated with its *idleness time* $I(y_i) \in \mathbb{R}_+$, a positive real number which is also referred to as *visitation frequency requirement*.

A perpetual movement schedule of two robots r_1, r_2 of speed 1 will be referred to as a *patrolling schedule* (robots may change movement direction instantaneously, and at no cost). Given a patrolling schedule \mathcal{A} , we define the *waiting time* $w_{\mathcal{A}}(y_i)$ of each point y_i as the supremum of the time difference between any two subsequent visitations by any of r_1, r_2 . When the patrolling schedule is clear from the context, we will drop the subscript in $w_{\mathcal{A}}$.

A patrolling schedule \mathcal{A} is called *feasible* if for all i , $w_{\mathcal{A}}(y_i) \leq I(y_i)$. Schedule \mathcal{A} is called *c-feasible*, or *c-approximation*, if $w_{\mathcal{A}}(y_i)/I(y_i) \leq c$, for each $i = 1, \dots, n$. Thus a feasible patrolling schedule is also 1-approximation, or 1-feasible.

An instance of PUF that admits a feasible patrolling schedule will be called *feasible*. In this paper we are concerned with the combinatorial optimization problem of minimizing the worst (normalized) violation of the idleness times for feasible instances, i.e., we are concerned with finding good approximation patrolling schedules, in which robots' trajectories can be determined efficiently in the size of the given input. We will call such patrolling schedules *efficient*.

The problem considered here is a close relative of *Pinwheel scheduling* [14] modeled by points with non-uniform deadlines (visitation-frequencies) spanned by a complete network with edges of uniform length. The complexity of Pinwheel scheduling depends on its representation. In particular we know that in the standard multi-set representation the problem is in NP, however, we still don't know whether it is NP-hard. One can try to get closer to this answer either by studying particular instances of the problem which can be decided [15] or instead by seeking approximate solutions [13]. In this paper we adopted the latter.

We use the following concepts in the analysis of our patrolling schedules. We associate each point y_i with its *range* defined as the closed intervals $R(y_i) = \left[\max \left\{ 0, y_i - \frac{I(y_i)}{2} \right\}, \min \left\{ 1, y_i + \frac{I(y_i)}{2} \right\} \right]$. Intuitively, $R(y_i)$ is the ball around y_i within which a robot can be moving introducing no violation to the visitation frequency requirement of y_i . We also group points y_i with respect to whether the extreme points fall within their range, i.e., we introduce:

$$S_{00} := \{y_i \in S : 0, 1 \notin R(y_i)\}, S_{01} := \{y_i \in S : 0 \notin R(y_i) \ni 1\}, \\ S_{10} := \{y_i \in S : 0 \in R(y_i) \not\ni 1\}, S_{11} := \{y_i \in S : 0, 1 \in R(y_i)\}.$$

3 Summary of Results & Paper Organization

Our main contribution pertains to efficient patrolling schedules (algorithms) of feasible PUF instances. In particular, the patrolling schedules we propose are highly efficient and simple, meaning that robots' trajectories can be determined by a few critical turning points, which can be computed in linear time in the number of points of the PUF instance. In order to do so, we provide in Section 4 some useful properties that all feasible PUF instances exhibit, and in particular a characterization of instances with “no problematic points”. For the latter instances, we also provide optimal feasible schedules (Theorem 1). Then we turn our attention to arbitrary feasible PUF instances. As a warm-up, we present in Section 5 a simple efficient 4-approximation patrolling schedule that does not require coordination between robots. Section 6 is devoted to the introduction of an elaborate and efficient $\sqrt{3}$ -approximation patrolling schedule. The execution of the patrolling schedule requires robots to remember at most two special turning points (that can be found efficiently), and, in some cases, their coordination so that they never come closer than a predetermined critical distance. Its performance analysis is based on further properties of feasible PUF instances that are presented in Section 6.1. In particular, the $\sqrt{3}$ -feasible patrolling schedule is the combination of Algorithms 1 and 2, presented in Sections 6.2 and 6.3 respectively, each of them performing well for a different spectrum of a special structural parameter of the given instance that we call *expansion*. Finally in Appendices A, B we also show that the analyses we provide for all our proposed algorithms are actually tight.

4 Characterization of (Some) Feasible PUF Instances

In this section we characterize feasible instances of PUF for which at least one of the extreme points falls within the range of each point.

Theorem 1. *An instance of PUF with $S_{00} = \emptyset$ is feasible if and only if the following conditions are satisfied:*

- (1) $S_{10} \subset \bigcap_{x \in S_{10}} R(x) = X_{10}$, and $0 \in X_{10}$.
- (2) $S_{01} \subset \bigcap_{x \in S_{01}} R(x) = X_{01}$, and $1 \in X_{01}$.

$$(3) \ S \subset [\bigcap_{x \in S_{10}} R(x)] \cup \bigcap_{x \in S_{01}} R(x) = X_{10} \cup X_{01}$$

Moreover, if conditions (1)-(3) are satisfied, then there exists an efficient 1-approximation partition-based patrolling schedule, i.e. a schedule in which every y_i is visited only by one robot.

In order to prove Theorem 1 we need few observations.

Observation 1 Assume \mathcal{A} is a feasible patrolling schedule. Then, for each $x \in S$ and each time window of length at least $\frac{I(x)}{2}$ during an execution of \mathcal{A} , at least one robot is in $R(x)$.

Proof. Reset time to $t_0 = 0$. Aiming at contradiction, assume there is no robot in $R(x)$ at $t \geq \frac{I(x)}{2}$. Since both robots have speed 1, no robot visited x in the period $[t - \frac{I(x)}{2}, t]$ and no robot is able to visit x in the period $[t, t + \frac{I(x)}{2}]$. Thus, \mathcal{A} is not a feasible patrolling schedule. \square

For simplicity, we may also assume that in any patrolling schedule (hence in feasible schedules as well), the position of robot r_1 in the unit interval is always to the left of the position of r_2 , as otherwise we can exchange the roles of the robots whenever they swap while they meet. We summarize as follows.

Observation 2 In any patrolling schedule of PUF, r_1 (r_2) is the only robot patrolling $y_1 = 0$ ($y_n = 1$), and r_1 stays always to the left of r_2 .

We are now ready to prove Theorem 1.

Proof (Theorem 1). First, we show implication (\Rightarrow) by contraposition. If Condition (1) is not satisfied, then there exists $x \in S_{10}$ such that $x \notin X_{10}$. Fix a feasible schedule \mathcal{A} . By Observation 2, we may assume that r_1 stays to the left of r_2 , throughout the execution of the schedule. By Observation 1, there must be a robot in X_{10} at each time t . Thus, r_1 must be in X_{10} at each time t . Consequently, $x \in S_{10} \setminus X_{10}$ is visited only by r_2 . But r_2 has to visit point $y_n = 1$, and by definition of S_{10} we know that $1 \notin R(x)$. Therefore, \mathcal{A} is not a feasible schedule. By definition of S_{10} , for all $x \in S_{10}$, we have $0 \in R(x)$. Therefore $0 \in X_{10}$. A similar argument proves that Condition (2) is satisfied.

By (1) and (2), there exist $a, b \in (0, 1)$ such that $X_{10} = [0, a]$ and $X_{01} = [b, 1]$. Now suppose that Condition (3) is not satisfied. Then $a < b$, and there is a point $x \in S$ such that $a < x < b$, and therefore neither r_1 nor r_2 can visit x .

For implication (\Leftarrow), assume that (1)-(3) are satisfied. Consider a partition traversal A , where r_1 is searching $X_{10} \setminus X_{01}$ and r_2 is searching X_{01} . Then, by the definition of the ranges $R(x)$, X_{10} and X_{01} , the traversal A is feasible. \square

The complication of instances when S_{00} is non empty is that in a feasible solution, points in S_{00} have to be interchangeably patrolled by both r_1, r_2 , which further requires appropriate synchronization between them. Even though a characterization of feasibility for such instances is eluding us, we provide below a necessary condition. This condition will be useful also later on.

Lemma 1. *For every feasible instance of PUF, we have $S_{00} \subset \bigcap_{x \in S} R(x)$.*

Proof. Suppose to the contrary, that there are $x \in S_{00}$ and $y \in S$, such that $x \notin R(y)$. By Observation 1, a robot is always present inside $R(y)$. Therefore the other robot must visit x . Without loss of generality assume that $y < x$. The robot that visits y cannot pass the point $y + \frac{I(y)}{2} < x$. Also the robot that visits x cannot pass the point $x + \frac{I(x)}{2}$. Since $x \in S_{00}$ then $x + \frac{I(x)}{2} < 1$. This means that no robot can visit point $y_n = 1$. \square

5 A Simple 4-Approximation Patrolling Schedule

In light of Theorem 1, it is interesting to study feasible instances of PUF that may have points that cannot be patrolled by one robot, i.e. for which $S_{00} \neq \emptyset$. As a warm-up, we provide a 4-feasible patrolling schedule for such instances. The advantage of this schedule is that robots' trajectories are simple and no coordination is required.

Theorem 2. *Feasible instances of PUF admit an 4-approximate patrolling schedule.*

Proof (Theorem 2). Let A be a feasible solution. Let $I = \min_{y \in S} I(y)$ and let $x \in S$ be such that $I(x) = I$. If $I \geq \frac{1}{2}$, then one robot patrolling the interval $[0, 1]$ gives a 4-approximation solution. Thus, we may assume that $I \leq \frac{1}{2}$.

According to Observation 1, at least one robot stays in $R(x)$ during A , at each time t . We claim that a nested traversal \mathcal{A} in which one robot traverses $[0, 1]$ and the other robot traverses $R(x)$ is a 4-approximation.

We split the interval $[0, 1]$ into $A = [0, a]$, $R(x) = [a, a + I] = [a, 1 - b]$ and $B = [1 - b, 1]$, where $a + I + b = 1$. First, note that the waiting time of each $y \in R(x)$ during \mathcal{A} is $w_{\mathcal{A}}(y) = 2I = 2I(x) \leq 2I(y)$. Thus, it remains to show that $w_{\mathcal{A}}(y) \leq 4I(y)$ for each point $y \in A \cup B$.

Without loss of generality assume that $|A| = a < b = |B|$. Using the assumption $I \leq \frac{1}{2}$ and $a + I + b = 1$, we have $a + b \geq \frac{1}{2}$, and therefore $b \geq \frac{1}{4}$. Using Observation 2, we consider a feasible schedule \mathcal{B} in which r_1 is always to the left of r_2 . By Observation 1, at least one robot stays in $R(x)$ at each time during \mathcal{B} . We consider the following cases:

- (**Case $y \in A$:**) As at each time moment there must exist a robot in $R(x)$, then in \mathcal{B} robot r_1 has to stay in $R(x)$ while r_2 is traversing $B = [1 - b, 1]$ twice to visit $y_n = 1$ and return to $R(x)$. Therefore the waiting time $w_{\mathcal{B}}$ satisfies $I(y) \geq w_{\mathcal{B}} \geq 2b \geq 2\frac{1}{4} = \frac{1}{2}$. On the other hand $w_{\mathcal{A}}(y) = 2 = 4\frac{1}{2} \leq 4I(y)$.
- (**Case $y \in B$:**) Let $y' = y - (a + I)$, thus y' is the distance of y to $R(x)$. Consider a time t during the execution of \mathcal{B} at which r_1 leaves $R(x)$ in order to visit the point 0. As r_2 must be in $R(x)$ at t , the last visit of y before t was at time $t' \leq t - y'$. Then, it has to stay in $R(x)$ for at least $2a + y'$. The time between two consecutive visits at y is at least $t + 2a + y' - (t - y') = 2a + 2y'$. On the other hand, in order to visit 1, r_2 has also time at least $2(1 - y')$ between two

consecutive visits of y . Altogether $w_{\mathcal{B}}(y) \geq \max\{2(a + y'), 2(b - y')\}$. Thus $w_{\mathcal{B}}(y) \geq \frac{1}{2}[2(a + y') + 2(b - y')] = a + b \geq \frac{1}{2}$. On the other hand $w_{\mathcal{A}}(y) = 2$ and thus $w_{\mathcal{A}}(y) \leq 4w_{\mathcal{B}} \leq 4I(y)$. \square

6 A $\sqrt{3}$ -Approximation Patrolling Schedule

The bottleneck toward patrolling instances of PUF is caused by points which require the coordination of both robots in order to be patrolled, i.e. instances in which $S_{00} \neq \emptyset$. In order to improve upon the 4-feasible schedule of Theorem 2, we need to understand better the visitation requirements of points in S_{00} , as well as their relative positioning in the path to be patrolled. The result of our analysis, and our main contribution, is an elaborate $\sqrt{3}$ -feasible patrolling schedule.

Theorem 3. *Feasible instances of PUF admit an efficient $\sqrt{3}$ -approximate patrolling schedule.*

In what follows, we explicitly assume that $S_{00} \neq \emptyset$, as otherwise, due to Theorem 1, we can easily find feasible schedules for instances of PUF that admit feasible solutions. Next, we introduce a key notion to our algorithms.

Definition 1. *Given an instance of PUF we identify critical points x_1, \dots, x_4 that are defined as follows: $\bigcap_{x \in S_{00}} R(x) = [x_1, x_4]$, and x_2, x_3 are the leftmost and rightmost points point in S_{00} , respectively. The instance is called α -expanding if $x_1 = \frac{\alpha}{1+\alpha}x_4$.*

Theorem 3 is an immediate corollary of the following Lemmata 2, 3 that we prove in subsequent Sections 6.2, 6.3, respectively. The lemmata are interesting in their own right, since they explicitly guarantee good approximate schedules as a function of the expansion of the given instance.

Lemma 2. *Feasible α -expanding instances of PUF admit an efficient $(1 + 2\alpha)$ -approximate patrolling schedule.*

Lemma 3. *Feasible α -expanding instances of PUF admit an efficient $\frac{2+\alpha}{1+\alpha}$ -approximate patrolling schedule.*

Lemmata 2, 3 above imply that any feasible α -expanding instance admits a $\min\left\{1 + 2\alpha, \frac{2+\alpha}{1+\alpha}\right\}$ feasible patrolling schedule. The achieved approximation is the worst when the instance is $\frac{\sqrt{3}-1}{2}$ -expanding, in which case, the patrolling schedule is $\sqrt{3}$ -feasible. This concludes the proof of Theorem 3.

Notably, our feasibility bounds above are tight. In Appendices A and B we show that for every α , there are feasible α -expanding PUF instances for which the performance of our patrolling schedules that prove Lemma 2 and Lemma 3 (see Sections 6.2, 6.3) is equal to the proposed bound. Hence, the performance analysis of our patrolling schedule showing Theorem 3 cannot be improved.

6.1 Useful Observations for Feasible PUF Instances

In an α -expanding instance of PUF we have that $x_1 = \alpha(x_4 - x_1)$. If the instance is also feasible, then by Lemma 1 we have that $S_{00} \subset \bigcap_{x \in S} R(x)$. Since $S_{00} \subset S$, we obtain that $S_{00} \subset \bigcap_{x \in S_{00}} R(x) = [x_1, x_4]$. Also, it is easy to see that for the critical points x_1, \dots, x_4 we have that $x_1 \leq x_2 < x_4$ and that $x_1 < x_3 \leq x_4$. In particular we may assume, without loss of generality, that $x_1 \leq 1 - x_4$, as otherwise we flip the order of all points. Also using Observation 2, we assume that the feasible schedule to the PUF instance has robot r_1 stay always to the left of r_2 .

Lemma 4. *Consider a feasible patrolling schedule \mathcal{A} for a PUF instance. Then*

- (1) *there is always a robot inside the interval $[x_1, x_4]$.*
- (2) *the interval $[0, x_1]$ is only traversed by r_1 and the interval $(x_4, 1]$ is only traversed by r_2 .*
- (3) *$0 \in R(x)$ for all $x \in [0, x_1)$, and $1 \in R(x)$ for all $x \in (x_4, 1]$.*
- (4) *$x_4 - x_3 \leq x_3 - x_1$ and $x_2 - x_1 \leq x_4 - x_1$.*

Proof. The proof of (1) is a direct consequence of Observation 1 and the fact that $[x_1, x_4]$ is the intersection of the ranges of all of the points of S_{00} .

During the execution of \mathcal{A} a robot needs to visit 0 and 1. Also, by (1) we know that there is always a robot inside $[x_1, x_4]$. Therefore while the robot r_2 is traversing $(x_4, 1]$ the robot r_1 has to stay inside $[x_1, x_4]$, and while robot r_1 is traversing $[0, x_1)$, the robot r_2 has to stay inside $[x_1, x_4]$. This implies that r_1 never passes x_4 and r_2 never passes x_1 . This proves (2). Part (3) follows directly from (2).

We now prove the first inequality of (4). Suppose to the contrary that $x_4 - x_3 > x_3 - x_1$, and thus $x_3 < \frac{x_1 + x_4}{2}$. For all $x \in S_{00}$ we have that $x_4 \in R(x)$. Therefore for all $x \in S_{00}$, $x_4 \leq x + \frac{I(x)}{2}$. Moreover x_3 is the rightmost point of S_{00} , hence $x \leq x_3 < \frac{x_1 + x_4}{2}$. Consequently $x_4 \leq x + \frac{I(x)}{2} \leq x_3 + \frac{I(x)}{2} < \frac{x_1 + x_4}{2} + \frac{I(x)}{2}$. This implies that $I(x) > \frac{x_4 - x_1}{2}$. So for all $x \in S_{00}$ we have $x - \frac{I(x)}{2} \leq x - \frac{x_4 - x_1}{2} < \frac{x_1 + x_4}{2} - \frac{x_4 - x_1}{2} = x_1$. Therefore there is a point $y \in (0, 1)$ such that for all $x \in S_{00}$, $x - \frac{I(x)}{2} \leq y < x_1$. Hence $y \in \bigcap_{x \in S_{00}} R(x)$ and $y < x_1$. This contradicts the fact that x_1 is the leftmost point of the intersection of the ranges of all the points of S_{00} . The proof of the second inequality of (4) follows by an analogous argument. \square

Lemma 5. *If there is a feasible solution for patrolling with two robots then the idle time of the points of S satisfy the following inequalities.*

$$I(x) \geq \begin{cases} \max\{2x, 2(1 - x - x_4 + x_1), x_4 - x_1\} & , x \in [0, x_1) \\ 2\max\{x_4 - x, x - x_1\} & , x \in [x_1, x_4] \\ \max\{2(1 - x), 2(x - x_4 + x_1), x_4 - x_1\} & , x \in (x_4, 1] \end{cases}$$

Proof. Let \mathcal{A} be a feasible solution and $x \in S$.

First assume that $x \in [0, x_1)$. By (2) of Lemma 4, in \mathcal{A} the points of $[0, x_1)$ are only visited by r_1 and $0 \in R(x)$. Thus, $I(x) \geq w_{\mathcal{A}}(x) \geq 2x$. Moreover robot

r_1 has to stay inside the interval $[x_1, x_4]$ for at least $2(1 - x_4)$ while the robot r_2 is traversing the interval $(x_4, 1]$ to visit 1. The time length for r_1 to traverse from x to x_1 , stay for at least $2(1 - x_4)$ inside $[x_1, x_4]$, and then traverse from x_1 to x is at least $2[(x_1 - x) + (1 - x_4)]$. Therefore, $I(x) \geq w_{\mathcal{A}(x)} \geq 2[(x_1 - x) + (1 - x_4)]$. On the other hand, by Lemma 1, we know that $x_3 \in R(x)$, and thus $\frac{I(x)}{2} \geq x_3 - x \geq x_3 - x_1$. By (3) of Lemma 4, $x_4 - x_3 \leq x_3 - x_1 \leq \frac{I(x)}{2}$. Therefore, $x_4 - x_1 = (x_4 - x_3) + (x_3 - x_1) \leq \frac{I(x)}{2} + \frac{I(x)}{2} = I(x)$. By the above discussion, and for all $x \in [0, x_1)$, we have $I(x) \geq \max\{2x, 2(1 - x - x_4 + x_1), x_4 - x_1\}$. A similar argument shows that for $x \in (x_4, 1]$ we have that $I(x) \geq \max\{2x, 2(1 - x - x_4 + x_1), x_4 - x_1\}$.

Now assume that $x \in [x_1, x_4]$. Then $x_1, x_4 \in R(x)$, and therefore $x - \frac{I(x)}{2} \leq x_1 \leq x_4 \leq x + \frac{I(x)}{2}$. This implies that $2(x - x_1) \leq I(x)$ and $2(x_4 - x) \leq I(x)$. So for all $x \in [x_1, x_4]$ we have $I(x) \geq \max\{x - x_1, x_4 - x\}$.

6.2 $(1 + 2\alpha)$ -Approximate Patrolling Schedules (Proof of Lemma 2)

Given a feasible α -expanding instance of PUF and using its critical points as in Definition 1, we propose the following algorithm.

Algorithm 1

- 1: Robot r_1 starts anywhere in $[0, x_3]$, and robot r_2 starts anywhere in $[x_3, 1]$.
 - 2: Repeat forever
 - 3: Robot r_1 zigzags inside $[0, x_3]$ and robot r_2 zigzags inside $[x_3, 1]$.
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Next we show that Algorithm 1 is $(1+2\alpha)$ -feasible, effectively proving Lemma 2. For this we analyze the waiting time $w(x)$ for all points $x \in S$.

Assume that $x \in [0, x_1)$. By Lemma 1, we know that $x_3 \in R(x)$. Moreover by (3) of Lemma 4, $0 \in R(x)$. Since r_1 zigzags inside $[0, x_3]$ then $w(x) \leq I(x)$.

Similarly, for $x \in (x_3, 1]$, by Lemma 1 and Lemma 4 we have $\{x_3, 1\} \subset R(x)$. Since r_2 zigzags inside $[x_3, 1]$ then $w(x) \leq I(x)$.

Finally, let $x \in [x_1, x_3]$. First assume that $x < x_3$. Then in Algorithm 1 the point x is only visited by r_1 . Since r_1 zigzags inside $[0, x_3]$ we have that $w(x) = 2\max\{x, x_3 - x\}$. We now compute the feasibility ratio. Clearly for the points $x \in [0, x_1) \cup (x_3, 1]$ we have that $\frac{w(x)}{I(x)} \leq 1$. So when $x \in [x_1, x_3]$, then by Lemma 5 $\frac{w(x)}{I(x)} \leq \frac{\max\{x, x_3 - x\}}{\max\{x - x_1, x_4 - x\}}$. First let $\max\{x - x_1, x_4 - x\} = x_4 - x$. Then $x \leq \frac{x_1 + x_4}{2}$. If $\max\{x, x_3 - x\} = x_3 - x$, as $x_3 \leq x_4$ we have that $\frac{w(x)}{I(x)} \leq 1$. If ,

on the other hand, $\max\{x, x_3 - x\} = x$, then we have

$$\begin{aligned}
\frac{w(x)}{I(x)} &\leq \frac{x}{x_4 - x} \leq \frac{\frac{x_1+x_4}{2}}{x_4 - \frac{x_1+x_4}{2}} \leq \frac{x_1 + x_4}{x_4 - x_1} \\
&= \frac{(x_4 - x_1) + 2x_1}{x_4 - x_1} = 1 + \frac{2x_1}{x_4 - x_1} \\
&= 1 + \frac{2x_1}{\frac{x_1}{\alpha}} \quad [\text{Using } x_1 = \alpha(x_4 - x_1)] \\
&= 1 + 2\alpha.
\end{aligned}$$

Now let $\max\{x - x_1, x_4 - x\} = x - x_1$. Then $x \geq \frac{x_1+x_4}{2}$. Moreover by (4) of Lemma 4, we have $x_3 \geq \frac{x_1+x_4}{2}$. Therefore $x_3 - x \leq x - x_1$. If $\max\{x, x_3 - x\} = x_3 - x$ then $\frac{w(x)}{I(x)} \leq 1$. So assume that $\max\{x, x_3 - x\} = x$, in which case

$$\begin{aligned}
\frac{w(x)}{I(x)} &\leq \frac{x}{x - x_1} \leq \frac{x - x_1 + x_1}{x - x_1} = 1 + \frac{x_1}{x - x_1} \\
&\leq 1 + \frac{x_1}{\frac{x_1+x_4}{2} - x_1} = 1 + \frac{2x_1}{x_4 - x_1} = 1 + \frac{2x_1}{\frac{x_1}{\alpha}} = 1 + 2\alpha.
\end{aligned}$$

Therefore, Algorithm 1 is a $(1 + 2\alpha)$ -approximation algorithm. Our analysis above is tight. For details, see Appendix A.

6.3 $\frac{2+\alpha}{1+\alpha}$ -Approximate Patrolling Schedules (Proof of Lemma 3)

The distributed algorithm that achieves feasibility performance $\frac{2+\alpha}{1+\alpha}$ is quite elaborate. At a high level, the two robots maintain some distance that never drops below a certain carefully chosen threshold. During the execution of the patrolling schedule, there will always be some robot patrolling (zigzagging within) a certain subinterval defined by critical points of the given instance. When the robots move towards each other, and their distance reaches the certain threshold, then robots exchange roles; the previously zigzagging robot abandons the subinterval and goes to patrol its extreme point, while the other robot starts zigzagging within the subinterval. The formal description of our algorithm follows. The reader may also consult Figure 1.

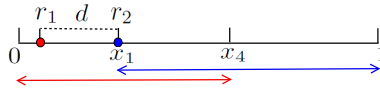


Fig. 1. The red arrow determines the patrolling area of r_1 and the blue arrow determines the patrolling area of r_2 .

Algorithm 2

- 1: Let $d = \frac{1}{1+\alpha} \min\{x_1, x_4 - x_1\}$.
 - 2: Robot r_1 starts at $x_1 - d$ and robot r_2 at x_1 .
 - 3: Repeat forever
 - 4: *Patrolling Schedule of r_1 :*
 - 5: **while** r_1 is inside the interval $[x_1, x_4]$ and the distance between the locations of r_1 and r_2 is at least d **do**
 - 6: Zigzag between points x_1 and x_4 .
 - 7: Visit point 0, then visit point x_1 , and then go to step 5.
 - 8: *Patrolling Schedule of r_2 :*
 - 9: **while** r_2 is inside the interval $[x_1, x_4]$ and the distance between the locations of r_2 and r_1 is at least d **do**
 - 10: Zigzag between points x_1 and x_4 .
 - 11: Visit point 1, then visit point x_4 , and then go to step 9.
-

Note that each robot has an explicit segment in which the points are visited by only that robot, *i.e.* $[0, x_1)$ is the explicit segment of r_1 and $(x_4, 1]$ is the explicit segment of r_2 . The trajectories of the robots overlap at $[x_1, x_4]$ where the points are visited by both r_1 and r_2 . The movements of the robots have two states: zigzagging inside $[x_1, x_4]$ and traversing their explicit segments twice. More precisely, once a robot enters $[x_1, x_4]$ it zigzags inside $[x_1, x_4]$ until the other robot is at distance d . Then it leaves $[x_1, x_4]$, traverses its explicit segment twice, and the same process repeats perpetually.

Note that by the definition of d , we know that $\min\{x_1, x_4 - x_1, 1 - x_4\} \geq d$. Therefore, the original placement of r_1 at $x_1 - d$ is compatible with Algorithm 2. The remaining of the section is devoted to proving that Algorithm 2 is $\frac{2+\alpha}{1+\alpha}$ -approximate, effectively proving Lemma 3. As a first step, we calculate the worst case waiting times $w(x)$ of all points in S .

Lemma 6. *The waiting times of points in S for Algorithm 2 are as follows.*

$$w(x) \begin{cases} = 2\max\{x, 1 - x - d\} & , x \in [0, x_1) \\ \leq 2\max\{x - x_1, x_4 - x\} + d & , x \in [x_1, x_4] \\ = 2\max\{1 - x, x - d\} & , x \in (x_4, 1] \end{cases}$$

Proof. Recall that $x_1 \leq 1 - x_4$, and in particular $\min\{x_1, x_4 - x_1, 1 - x_4\} \geq d$.

Case $0 \leq x < x_1$: Point x is only visited by robot r_1 . We now calculate the time interval between two consecutive visitations of x by r_1 . We distinguish two subcases.

First consider the subcase where r_1 is moving to the left when it visits x . Before r_1 visits x again, it has to visit 0 and then return to x . Therefore, the time between the two visitations of x is $2x$.

Second consider the subcase in which r_1 is moving to the right when it visits x . Before r_1 visits x again, it has to visit x_1 (*i.e.* enter interval $[x_1, x_4]$), zigzag between points x_1 and x_4 until its distance to the other robots becomes d , and then r_1 exits the interval $[x_1, x_4]$ and return to x . Below we compute the total time between these two visitations of x by r_1 .

- (1a): r_1 traverses from x to x_1 : it takes $x_1 - x$.
- (1b): r_1 zigzags inside $[x_1, x_4]$: at the time that r_1 arrives at x_1 and starts zigzagging inside $[x_1, x_4]$, robot r_2 is at distance d from r_1 and it is moving to the right to visit 1 and return. Also, at the time that r_1 arrives at x_1 to exit the interval $[x_1, x_4]$, the distance between r_1 and r_2 is d and robot r_2 is moving to the left to zigzag inside the interval $[x_1, x_4]$. Therefore, the time r_1 spends inside the interval $[x_1, x_4]$ is equal to the time that r_2 spends to traverse from $x_1 + d$ to 1 and return to $x_1 + d$, which is $2(1 - x_1 - d)$.
- (1c): r_1 traverses from x_1 to x : it takes $x_1 - x$.
- Using (1a,1b,1c) above, we conclude that the total time between two consecutive visitations of x by r_1 is $2(1 - x - d)$.
- Taking into consideration both subcases, the overall (worst case) waiting time of x is $2\max\{x, 1 - x - d\}$.

Case $x_4 < x \leq 1$: The analysis is analogous to the previous case.

Case $x_1 \leq x \leq x_4$: Point x is visited by both r_1 and r_2 . We consider two subcases

- (1) The two consecutive visits of x are by the same robot r_1 or r_2 : this case occurs when either of r_1 or r_2 zigzags inside the interval $[x_1, x_4]$. Therefore $w(x) = 2\max\{x_4 - x, x - x_1\}$.
- (2) The two consecutive visits of x are by different robots r_1 and r_2 : this case occurs when one robot is exiting the interval $[x_1, x_4]$ and the other one is entering it.

First suppose that r_1 visits x and the next visit of x is performed by r_2 . The worst waiting time in this case occurs when r_1 is about to visit x but the distance between r_1 and r_2 reduces to d and so r_1 turns away from x . Then r_2 visits x after at most d time steps. Note that since $x_1 \geq d$ the visit of x by r_2 is guaranteed. Therefore $w(x) \leq 2(x - x_1) + d$. Now assume that r_2 visits x and the next visit of x is performed by r_1 . By a similar discussion we have that $w(x) \leq 2(x_4 - x) + d$. This implies that $w(x) \leq 2\max\{x - x_1, x_4 - x\} + d$.

By Subcases 1,2 above we conclude that $w(x) \leq 2\max\{x - x_1, x_4 - x\} + d$, for all $x \in [x_1, x_4]$.

□

The proof of Lemma 3 follows by upper bounding $\max_{x \in S} \left\{ \frac{w(x)}{I(x)} \right\}$. Using Lemma 5 and Lemma 6, we see that the approximation ratio of Algorithm 2 is no more than

$$\frac{w(x)}{I(x)} \leq \begin{cases} \frac{2\max\{x, 1-x-d\}}{\max\{2x, 2(1-x-x_4+x_1), x_4-x_1\}} & , x \in [0, x_1] \\ \frac{2\max\{x-x_1, x_4-x\}+d}{2\max\{x_4-x, x-x_1\}} & , x \in [x_1, x_4] \\ \frac{2\max\{1-x, x-d\}}{\max\{2(1-x), 2(x-x_4+x_1), x_4-x_1\}} & , x \in (x_4, 1] \end{cases} \quad (1)$$

Using that $d = \frac{\min\{x_1, x_4-x_1\}}{1+\alpha}$, and the fact that the given instance is α -expanding, i.e. that $x_1 = \alpha(x_4 - x_1)$, and after some tedious and purely algebraic calculations, we see that $\frac{w(x)}{I(x)} \leq \frac{2+\alpha}{1+\alpha}$ for all $x \in S$, as wanted. Details can be found in Appendix C.

7 Conclusion

The paper investigated the problem of patrolling a line segment by two robots when time-patrolling constraints are placed on the frequency of visitation of all the points of the line. As shown in this study, finding “efficient” trajectories that satisfy the requirements or even deciding on their existence for two robots turns out to be a highly intricate problem. Nothing better is known aside from the $\sqrt{3}$ -approximation algorithm for two robots on a line presented in this work. The patrolling problem with constraints is also open for more general graph topologies (e.g, cycles, trees, etc.). Further, the case of patrolling with constraints for multiple robots is completely unexplored in all topologies, including for the line segment.

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A Tightness Analysis of Algorithm 1

For every α , we provide a feasible α -expanding PUF instance for which the performance of Algorithm 1 is exactly equal to $1 + 2\alpha$. The instance is as follows. Choose $x_1 \in (0, 1)$ such that $2x_1 + \frac{x_1}{\alpha} = 1$. Consider three points $y_0 = 0$, $y_1 = x_1(1 + \frac{1}{2\alpha})$, and $y_2 = 1$. Let $I(y_0) = I(y_2) = 4x_1 + \frac{x_1}{\alpha}$ and $I(y_1) = \frac{x_1}{\alpha}$.

Let \mathcal{A} be a solution as follows. At the beginning, the robot r_1 locates at x_1 and the robot r_2 locates at y_1 , and both robots move to the right. See Figure 2. The patrolling segment of r_1 is $[y_0, y_1]$ and the patrolling segment of r_2 is

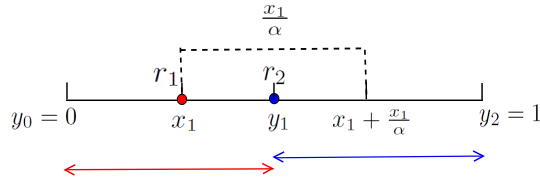


Fig. 2. The red arrow determines the patrolling area of r_1 in \mathcal{A} and the blue arrow determines the patrolling area of r_2 in \mathcal{A} .

$[y_1, y_2]$. The robot r_1 moves back and forth inside the interval $[y_0, y_1]$ and each time r_1 visits y_1 stays at y_1 for $2x_1$. Similarly the robot r_2 moves back and forth inside the interval $[y_1, y_2]$ and each time it visits y_1 stays at y_1 for $2x_1$. Then $w_{\mathcal{A}}(y_0) = w_{\mathcal{A}}(y_2) = 4x_1 + \frac{x_1}{\alpha} = I(y_0) = I(y_2)$ and $w_{\mathcal{A}}(y_1) = \frac{x_1}{2\alpha} < I(y_1)$. Therefore \mathcal{A} is a feasible solution.

Now consider Algorithm 1. For the above example we have $x_2 = x_3 = y_1$. Therefore r_1 zigzags inside $[0, y_1]$ and r_2 zigzags inside $[y_1, 1]$. Since the movement of the robots is not coordinated in Algorithm 1 the worst waiting time of y_1 in Algorithm 1 is $2x_1 + \frac{x_1}{\alpha}$. This implies that

$$\frac{w(x)}{I(x)} = \frac{x_1(2 + \frac{1}{\alpha})}{\frac{x_1}{\alpha}} = 1 + 2\alpha.$$

B Tightness Analysis of Algorithm 2

For every α , we provide a feasible α -expanding PUF instance for which the performance of Algorithm 2 is exactly equal to $\frac{2+\alpha}{1+\alpha}$. For this we consider two cases.

Case 1. $\alpha < 1$: Choose $x_1 \in (0, 1)$ such that $x_1 + 2\frac{x_1}{\alpha} = 1$. Consider four points $y_0 = 0$, $y_1 = x_1 + \frac{x_1}{2\alpha}$, $y_2 = \frac{2x_1}{\alpha}$, and $y_3 = 1$. See Figure 3. Let $I(y_0) = I(y_3) = x_1(2 + \frac{3}{\alpha})$, $I(y_1) = \frac{x_1}{\alpha}$, and $I(y_2) = \frac{2x_1}{\alpha}$.

Let \mathcal{A} be a solution as follows.

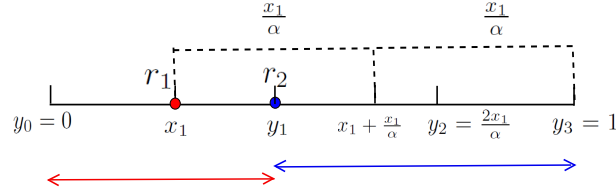


Fig. 3. The red arrow determines the patrolling area of r_1 in \mathcal{A} and the blue arrow determines the patrolling area of r_2 in \mathcal{A} .

- (1) At the beginning, the robot r_1 locates at x_1 and the robot r_2 locates at y_1 , and both robots move to the right.
- (2) The patrolling segment of r_1 is $[y_0, y_1]$ and the patrolling segment of r_2 is $[y_1, y_3]$.
- (3) The robot r_1 moves back and forth inside the interval $[y_0, y_1]$ and each time r_1 visits y_1 stays at y_1 until r_2 is at point $y_1 + \frac{x_1}{\alpha}$ and is moving to the left. Then r_1 moves to the left, visits 0 and returns to y_1 .
- (4) The robot r_2 moves back and forth inside the interval $[y_1, y_3]$ and each time it visits y_1 stays at y_1 for $2x_1$. Then r_2 moves to the right, visits y_2 and 1 and returns to y_1 .

First we analyze the movement of the robots in \mathcal{A} . The robot r_1 leaves y_1 when the distance between r_1 and r_2 is $\frac{x_1}{\alpha}$. Note that this is possible since the length of $[y_1, y_3]$ is greater than $\frac{x_1}{\alpha}$. The next visit of r_1 from y_1 occurs after

$$\frac{x_1}{2\alpha} + 2x_1 + \frac{x_1}{2\alpha} = \frac{x_1}{\alpha} + 2x_1,$$

which is the time that r_1 spends to visit 0 and return to y_1 . During the time $\frac{x_1}{\alpha} + 2x_1$ the robots r_1 traverse from $y_1 + \frac{x_1}{\alpha}$ to y_1 and stays there for $2x_1$. Therefore by the time r_1 arrives at y_1 , the robot r_2 leaves y_1 . We now compute the waiting time of \mathcal{A} .

The point y_0 is only visited by r_1 . So the waiting time of 0 is equal “two times the length of the interval $[0, y_1]$ plus the time r_1 stops at y_1 ”. The stop time of r_1 at y_1 is equal the time r_2 traverse from y_1 to 1 and return to the point $y_1 + \frac{x_1}{\alpha}$ which is $\frac{2x_1}{\alpha}$. Therefore

$$w_{\mathcal{A}}(y_0) = \frac{2x_1}{\alpha} + 2x_1 + \frac{x_1}{\alpha} = x_1(2 + \frac{3}{\alpha}) = I(x).$$

The point y_1 is visited by both r_1, r_2 . First suppose that r_1 is waiting at point y_1 . The robot r_1 leaves y_1 when r_2 is at distance $\frac{x_1}{\alpha}$ of y_1 . So the next visit of y_1 occurs after $\frac{x_1}{\alpha}$ time by r_2 . Also, as we discussed above by the time r_2 is ready to leave y_1 the robot r_1 arrives at y_1 . So in this case there is no time interval between the visits of r_2 and r_1 . Therefore

$$w_{\mathcal{A}}(y_1) = \frac{x_1}{\alpha} = I(x).$$

The point y_2 is only visited by r_2 . If r_2 is moving to the left when it visits y_2 then the next visit of y_2 occurs in $\frac{2x_1}{\alpha}$. This is the time r_2 spends to visit y_1 and stays at it for $2x_1$ and returns to y_2 . If r_2 is moving to the right when it visits y_2 then clearly the next visit of y_2 occurs in $2(1 - \frac{2x_1}{\alpha}) = 2x_1$. Therefore

$$w_{\mathcal{A}}(y_2) = \max\{2x_1, 2\frac{x_1}{\alpha}\} = \frac{2x_1}{\alpha} = I(x).$$

For $y_3 = 1$ it is easy to see that

$$w_{\mathcal{A}}(y_3) = 2(1 - y_1) + 2x_1 = 2(1 - x_1 - \frac{x_1}{2\alpha}) + 2x_1 = 2 - \frac{x_1}{\alpha} = x_1(2 + \frac{3}{\alpha}).$$

The last equation follows from the fact that $x_1 + \frac{2x_1}{\alpha} = 1$. Therefore \mathcal{A} is a feasible solution.

Now consider Algorithm 2. For the above example we have $d = \frac{x_1}{1+\alpha}$, $x_2 = x_3 = y_1$, and $x_4 = x_1 + \frac{x_1}{\alpha} < \frac{2x_1}{\alpha}$. Therefore by Lemma 6

$$w(y_2) = 2\max\{x_1, \frac{2x_1}{\alpha} - \frac{x_1}{1+\alpha}\} = \frac{2x_1}{\alpha} - \frac{x_1}{1+\alpha}.$$

Therefore

$$\frac{w(y_2)}{I(y_2)} = \frac{\frac{2x_1}{\alpha} - \frac{x_1}{1+\alpha}}{\frac{x_1}{\alpha}} = 1 + \frac{x_1(\frac{1}{1+\alpha})}{\frac{x_1}{\alpha}} = \frac{2+\alpha}{1+\alpha}.$$

Case 2. $\alpha \geq 1$: Choose $x_1 \in (0, 1)$ such that $2x_1 + \frac{x_1}{\alpha} = 1$ and let $0 < \epsilon < x_1$. Consider four points $y_0 = 0$, $y_1 = x_1 - \epsilon$, $y_2 = x_1 + \frac{x_1}{2\alpha}$, and $y_3 = 1$. See Figure 4. Let $I(y_0) = I(y_3) = 2(1 - x_1)$, $I(y_1) = 2(x_1 + \epsilon)$, and $I(y_2) = \frac{x_1}{\alpha}$.

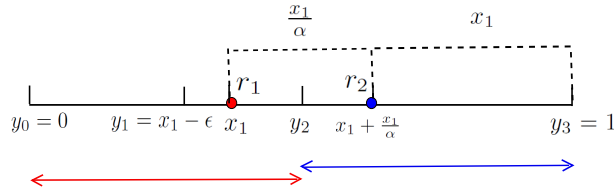


Fig. 4. The red arrow determines the patrolling area of r_1 in \mathcal{A} and the blue arrow determines the patrolling area of r_2 in \mathcal{A} .

Let \mathcal{A} be a solution as follows.

- (1) At the beginning, the robot r_1 locates at x_1 and the robot r_2 locates at $x_1 + \frac{x_1}{\alpha}$, and both robots move to the right.
- (2) The patrolling segment of r_1 is $[y_0, y_2]$ and the patrolling segment of r_2 is $[y_2, y_3]$.

- (3) The robot r_1 moves back and forth inside the interval $[y_0, y_2]$ and each time r_1 visits y_1 stays at y_1 for $\frac{x_1}{\alpha}$. Then r_1 moves to the left, visits y_1 and 0 and returns to y_1 .
- (4) The robot r_2 moves back and forth inside the interval $[y_2, y_3]$ and each time it visits y_1 stays at y_1 for $\frac{x_1}{\alpha}$. Then r_2 moves to the right, visits 1 and returns to y_1 .

First we analyze the movement of the robots in \mathcal{A} . The robot r_1 leaves y_1 when the distance between r_1 and r_2 is $\frac{x_1}{\alpha}$, and similarly the robot r_2 leaves y_1 when the distance between r_1 and r_2 is $\frac{x_1}{\alpha}$. This is possible since the length of the intervals $[0, y_2]$ and $[y_2, 1]$ is $x_1 + \frac{x_1}{2\alpha}$ and $x_1 \geq \frac{x_1}{\alpha}$. It is easy to see that the

$$w_{\mathcal{A}}(y_0) = w_{\mathcal{A}}(y_3) = 2(1 - x_1).$$

The point $y_1 = x_1 - \epsilon$ is only visited by r_1 and so

$$\begin{aligned} w(y_1) &= 2\max\{y_1, y_2 - y_1 + \frac{x_1}{2\alpha}\} \\ &= 2\max\{x_1 - \epsilon, \frac{x_1}{\alpha} + \epsilon\} \leq 2(x_1 + \epsilon). \end{aligned}$$

In \mathcal{A} the point $y_2 = x_1 + \frac{x_1}{2\alpha}$ is patrolled by both r_1 and r_2 in such a way that one robot stays at y_2 until the other robot is moving towards y_2 and is at distance $\frac{x_1}{\alpha}$ from y_2 . This implies that $w_{\mathcal{A}}(y_2) \leq \frac{x_1}{\alpha} = I(x)$.

From the above discussion we have that \mathcal{A} is a feasible solution. Now consider Algorithm 2. For the above example we have $d = \frac{x_1}{\alpha(1+\alpha)}$, $x_2 = x_3 = y_2$. Since $y_1 = x_1 - \epsilon < x_1$ by Lemma 6 we have that

$$\begin{aligned} w(y_1) &= 2\max\{x_1 - \epsilon, 1 - (x_1 - \epsilon) - \frac{x_1}{\alpha(1+\alpha)}\} \\ &= 2\max\{x_1 - \epsilon, x_1 + \epsilon + x_1(\frac{1}{1+\alpha})\} = x_1 + \epsilon + x_1(\frac{1}{1+\alpha}). \end{aligned}$$

In the above equation we use the facts that $2x_1 + \frac{x_1}{\alpha} = 1$ and $\alpha \geq 1$. Therefore

$$\begin{aligned} \frac{w(y_1)}{I(y_1)} &= \frac{x_1 + \epsilon + x_1(\frac{1}{1+\alpha})}{x_1 + \epsilon} \\ &= 1 + \frac{1}{1+\alpha} \frac{x_1}{x_1 + \epsilon}. \end{aligned}$$

This implies that as ϵ converges to 0 the feasibility ratio of Algorithm 2 converges to $1 + \frac{1}{1+\alpha} = \frac{2+\alpha}{1+\alpha}$.

C Omitted Proofs of Section 6.3

Using that $d = \frac{\min\{x_1, x_4 - x_1\}}{1+\alpha}$, and the fact that the given instance is α -expanding, i.e. that $x_1 = \alpha(x_4 - x_1)$, we prove that $w(x)/I(x)$, as it reads in (1), is upper bounded by $\frac{2+\alpha}{1+\alpha}$. Naturally, we distinguish cases for the location of x with respect to the three subintervals $[1, x_1]$, $[x_1, x_4]$, $(x_4, 1 - x_4]$. For the sake of exposition, we split the proof in 3 corresponding Lemmata.

Lemma 7. For all $0 \leq x < x_1$ we have $\frac{w(x)}{I(x)} \leq \frac{2+\alpha}{1+\alpha}$.

Proof. Using (1) we have that

$$\frac{w(x)}{I(x)} \leq \frac{\max\{x, 1-x-d\}}{\max\{x, 1-x-x_4+x_1, x_4-x_1\}}.$$

If $x > \frac{1-d}{2}$ then $\max\{x, 1-x-d\} = x$, and so $\frac{w(x)}{I(x)} \leq 1$. So assume that $x \leq \frac{1-d}{2}$. We consider two cases:

Case 1: $\alpha \leq 1$. Then $x_1 = \alpha(x_4 - x_1) \leq x_4 - x_1$, and so $d = \frac{1}{1+\alpha}x_1$.

$$\frac{w(x)}{I(x)} \leq \frac{1-x-\frac{x_1}{1+\alpha}}{\max\{x, 1-x-\frac{x_1}{\alpha}, \frac{x_1}{\alpha}\}} \leq \frac{1-x-\frac{x_1}{1+\alpha}}{\max\{1-x-\frac{x_1}{\alpha}, \frac{x_1}{\alpha}\}}.$$

First let $\max\{1-x-\frac{x_1}{\alpha}, \frac{x_1}{\alpha}\} = \frac{x_1}{\alpha}$. Then $\frac{x_1}{\alpha} \geq 1-x-\frac{x_1}{\alpha}$, which implies that $x \geq 1-\frac{2x_1}{\alpha}$. Therefore

$$\begin{aligned} \frac{w(x)}{I(x)} &\leq \frac{1-x-\frac{x_1}{1+\alpha}}{\frac{x_1}{\alpha}} \leq \frac{1-(1-\frac{2x_1}{\alpha})-\frac{x_1}{1+\alpha}}{\frac{x_1}{\alpha}} \\ &= \frac{\frac{2x_1}{\alpha}-\frac{x_1}{1+\alpha}}{\frac{x_1}{\alpha}} = 1 + \frac{x_1(\frac{1}{\alpha}-\frac{1}{1+\alpha})}{\frac{x_1}{\alpha}} \\ &= 1 + \frac{1}{1+\alpha} = \frac{2+\alpha}{1+\alpha}. \end{aligned}$$

Now let $\max\{1-x-\frac{x_1}{\alpha}, \frac{x_1}{\alpha}\} = 1-x-\frac{x_1}{\alpha}$. Then $\frac{x_1}{\alpha} \leq 1-x-\frac{x_1}{\alpha}$, which implies that $x \leq 1-\frac{2x_1}{\alpha}$.

$$\begin{aligned} \frac{w(x)}{I(x)} &\leq \frac{1-x-\frac{x_1}{1+\alpha}}{1-x-\frac{x_1}{\alpha}} = \frac{1-x-\frac{x_1}{\alpha}+\frac{x_1}{\alpha}-\frac{x_1}{1+\alpha}}{1-x-\frac{x_1}{\alpha}} \\ &= 1 + \frac{\frac{x_1}{\alpha(1+\alpha)}}{1-x-\frac{x_1}{\alpha}} \leq 1 + \frac{\frac{x_1}{\alpha(1+\alpha)}}{1-(1-\frac{2x_1}{\alpha})-\frac{x_1}{\alpha}} \\ &= 1 + \frac{1}{1+\alpha} = \frac{2+\alpha}{1+\alpha}. \end{aligned}$$

Case 2: $\alpha \geq 1$. Then $x_1 = \alpha(x_4 - x_1) \geq x_4 - x_1$, and $d = \frac{1}{1+\alpha}(x_4 - x_1) = \frac{1}{\alpha(1+\alpha)}x_1$.

$$\frac{w(x)}{I(x)} \leq \frac{1-x-\frac{x_1}{\alpha(1+\alpha)}}{\max\{x, 1-x-\frac{x_1}{\alpha}, \frac{x_1}{\alpha}\}} \leq \frac{1-x-\frac{x_1}{\alpha(1+\alpha)}}{\max\{x, 1-x-\frac{x_1}{\alpha}\}}.$$

First let $\max\{x, 1-x-\frac{x_1}{\alpha}\} = x$. Then $x \geq 1-x-\frac{x_1}{\alpha}$, which implies that $x \geq \frac{1}{2} - \frac{x_1}{2\alpha}$.

$$\begin{aligned} \frac{w(x)}{I(x)} &\leq \frac{1-x-\frac{x_1}{\alpha(1+\alpha)}}{x} \leq \frac{1-(\frac{1}{2}-\frac{x_1}{2\alpha})-\frac{x_1}{\alpha(1+\alpha)}}{\frac{1}{2}-\frac{x_1}{2\alpha}} \\ &= \frac{\frac{1}{2}-\frac{x_1}{2\alpha}+(\frac{x_1}{\alpha}-\frac{x_1}{\alpha(1+\alpha)})}{\frac{1}{2}-\frac{x_1}{2\alpha}} = 1 + \frac{x_1(\frac{1}{1+\alpha})}{\frac{1}{2}-\frac{x_1}{2\alpha}}. \end{aligned}$$

By assumption we know that $x_1 \leq 1 - x_4$, and $x_4 - x_1 = \frac{x_1}{\alpha}$. Moreover $x_1 + (x_4 - x_1) + (1 - x_4) = 1$. Therefore $x_1 + \frac{x_1}{\alpha} + x_1 \leq 1$, which implies that $x_1 \leq \frac{\alpha}{2\alpha+1}$. Therefore

$$\begin{aligned} 1 + \frac{x_1(\frac{1}{1+\alpha})}{\frac{1}{2} - \frac{x_1}{2\alpha}} &\leq 1 + \frac{\frac{\alpha}{2\alpha+1}(\frac{1}{1+\alpha})}{\frac{1}{2} - \frac{1}{2(2\alpha+1)}} = 1 + \frac{\frac{\alpha}{2\alpha+1}(\frac{1}{1+\alpha})}{\frac{\alpha}{2\alpha+1}} \\ &= 1 + \frac{1}{1+\alpha} = \frac{2+\alpha}{1+\alpha}. \end{aligned}$$

Now let $\max\{x, 1 - x - \frac{x_1}{\alpha}\} = 1 - x - \frac{x_1}{\alpha}$. Then $x \leq 1 - x - \frac{x_1}{\alpha}$, which implies that $x \geq 1 - \frac{2x_1}{\alpha}$. In the following calculations we use the fact that $x \leq x_1$.

$$\begin{aligned} \frac{w(x)}{I(x)} &\leq \frac{1 - x - \frac{x_1}{\alpha(1+\alpha)}}{1 - x - \frac{x_1}{\alpha}} = \frac{1 - x - \frac{x_1}{\alpha} + \frac{x_1}{\alpha} - \frac{x_1}{\alpha(1+\alpha)}}{1 - x - \frac{x_1}{\alpha}} \\ &= 1 + \frac{x_1(\frac{1}{\alpha} - \frac{1}{\alpha(1+\alpha)})}{1 - x - \frac{x_1}{\alpha}} \leq 1 + \frac{x_1(\frac{1}{\alpha} - \frac{1}{\alpha(1+\alpha)})}{1 - x_1 - \frac{x_1}{\alpha}} \\ &\leq 1 + \frac{\frac{\alpha}{2\alpha+1}(\frac{1}{1+\alpha})}{1 - \frac{\alpha}{2\alpha+1}(1 + \frac{1}{\alpha})} = 1 + \frac{\frac{\alpha}{2\alpha+1}(\frac{1}{1+\alpha})}{1 - \frac{1+\alpha}{2\alpha+1}} \\ &= 1 + \frac{\frac{\alpha}{2\alpha+1}(\frac{1}{1+\alpha})}{\frac{\alpha}{2\alpha+1}} = 1 + \frac{1}{1+\alpha} = \frac{2+\alpha}{1+\alpha}. \end{aligned}$$

This proves that for all $x \in [0, x_1]$ the ratio $\frac{w(x)}{I(x)}$ is at most $\frac{2+\alpha}{1+\alpha}$. \square

Lemma 8. For all $x_1 \leq x \leq x_4$ we have $\frac{w(x)}{I(x)} \leq \frac{2+\alpha}{1+\alpha}$.

Proof. Using (1) we have that

$$\frac{w(x)}{I(x)} \leq \frac{\max\{x - x_1, x_4 - x\} + \frac{d}{2}}{\max\{x - x_1, x_4 - x\}} = 1 + \frac{d}{\max\{x - x_1, x_4 - x\}}.$$

If $x - x_1 \leq x_4 - x$ then $x \leq \frac{x_1+x_4}{2}$ and $\max\{x - x_1, x_4 - x\} = x_4 - x$. Moreover $x_4 - x \geq x_4 - \frac{x_1+x_4}{2} = x_4 - x_1$. Therefore

$$\frac{w(x)}{I(x)} \leq 1 + \frac{d}{x_4 - x_1}.$$

Since $d \leq \frac{1}{1+\alpha}(x_4 - x_1)$ we have

$$\frac{w(x)}{I(x)} \leq 1 + \frac{1}{1+\alpha} = \frac{2+\alpha}{1+\alpha}.$$

Now let $x_4 - x \leq x - x_1$. Then $x \geq \frac{x_1+x_4}{2}$ and $\max\{x - x_1, x_4 - x\} = x - x_1$. Moreover, $x - x_1 \geq \frac{x_1+x_4}{2} - x_1 = x_4 - x_1$, and thus

$$\frac{w(x)}{I(x)} \leq 1 + \frac{d}{x_4 - x_1} \leq 1 + \frac{1}{1+\alpha} = \frac{2+\alpha}{1+\alpha}.$$

\square

Lemma 9. For all $x_4 < x \leq 1$ we have $\frac{w(x)}{I(x)} \leq \frac{2+\alpha}{1+\alpha}$.

Proof. Using (1) we have that

$$\frac{w(x)}{I(x)} \leq \frac{\max\{1-x, x-d\}}{\max\{1-x, x-x_4+x_1, x_4-x_1\}}.$$

If $x > \frac{1-d}{2}$ then $\max\{1-x, x-d\} = 1-x$, and so $\frac{w(x)}{I(x)} \leq 1$. So assume that $x \leq \frac{1-d}{2}$. We consider two cases:

Case 1: $\alpha \leq 1$. Then $x_1 = \alpha(x_4 - x_1) \leq x_4 - x_1$, and so $d = \frac{1}{1+\alpha}x_1$.

$$\frac{w(x)}{I(x)} \leq \frac{x - \frac{x_1}{1+\alpha}}{\max\{1-x, x - \frac{x_1}{\alpha}, \frac{x_1}{\alpha}\}} \leq \frac{x - \frac{x_1}{1+\alpha}}{\max\{x - \frac{x_1}{\alpha}, \frac{x_1}{\alpha}\}}.$$

First let $\max\{x - \frac{x_1}{\alpha}, \frac{x_1}{\alpha}\} = \frac{x_1}{\alpha}$. Then $\frac{x_1}{\alpha} \geq x - \frac{x_1}{\alpha}$, which implies that $x \leq \frac{2x_1}{\alpha}$. Therefore

$$\begin{aligned} \frac{w(x)}{I(x)} &\leq \frac{x - \frac{x_1}{1+\alpha}}{\frac{x_1}{\alpha}} \leq \frac{\frac{2x_1}{\alpha} - \frac{x_1}{1+\alpha}}{\frac{x_1}{\alpha}} \\ &= \frac{\frac{2x_1}{\alpha} - \frac{x_1}{1+\alpha}}{\frac{x_1}{\alpha}} = 1 + \frac{x_1(\frac{1}{\alpha} - \frac{1}{1+\alpha})}{\frac{x_1}{\alpha}} \\ &= 1 + \frac{1}{1+\alpha} = \frac{2+\alpha}{1+\alpha}. \end{aligned}$$

Now let $\max\{x - \frac{x_1}{\alpha}, \frac{x_1}{\alpha}\} = x - \frac{x_1}{\alpha}$. Then $\frac{x_1}{\alpha} \leq x - \frac{x_1}{\alpha}$, which implies that $x \geq \frac{2x_1}{\alpha}$.

$$\begin{aligned} \frac{w(x)}{I(x)} &\leq \frac{x - \frac{x_1}{1+\alpha}}{x - \frac{x_1}{\alpha}} = \frac{x - \frac{x_1}{\alpha} + \frac{x_1}{\alpha} - \frac{x_1}{1+\alpha}}{x - \frac{x_1}{\alpha}} \\ &= 1 + \frac{\frac{x_1}{\alpha(1+\alpha)}}{x - \frac{x_1}{\alpha}} \leq 1 + \frac{\frac{x_1}{\alpha(1+\alpha)}}{\frac{2x_1}{\alpha} - \frac{x_1}{\alpha}} \\ &= 1 + \frac{1}{1+\alpha} = \frac{2+\alpha}{1+\alpha}. \end{aligned}$$

Case 2: $\alpha \geq 1$. Then $x_1 = \alpha(x_4 - x_1) \geq x_4 - x_1$, and $d = \frac{1}{1+\alpha}(x_4 - x_1) = \frac{1}{\alpha(1+\alpha)}x_1$.

$$\frac{w(x)}{I(x)} \leq \frac{x - \frac{x_1}{\alpha(1+\alpha)}}{\max\{1-x, x - \frac{x_1}{\alpha}, \frac{x_1}{\alpha}\}} \leq \frac{x - \frac{x_1}{\alpha(1+\alpha)}}{\max\{1-x, x - \frac{x_1}{\alpha}\}}.$$

First let $\max\{1-x, x - \frac{x_1}{\alpha}\} = 1-x$. Then $1-x \geq x - \frac{x_1}{\alpha}$, which implies that $x \leq \frac{1}{2} + \frac{x_1}{2\alpha}$.

$$\begin{aligned} \frac{w(x)}{I(x)} &\leq \frac{x - \frac{x_1}{\alpha(1+\alpha)}}{1-x} \leq \frac{\frac{1}{2} + \frac{x_1}{2\alpha} - \frac{x_1}{\alpha(1+\alpha)}}{\frac{1}{2} - \frac{x_1}{2\alpha}} \\ &= \frac{\frac{1}{2} - \frac{x_1}{2\alpha} + (\frac{x_1}{\alpha} - \frac{x_1}{\alpha(1+\alpha)})}{\frac{1}{2} - \frac{x_1}{2\alpha}} = 1 + \frac{x_1(\frac{1}{1+\alpha})}{\frac{1}{2} - \frac{x_1}{2\alpha}}. \end{aligned}$$

By assumption we know that $x_1 \leq 1 - x_4$, and $x_4 - x_1 = \frac{x_1}{\alpha}$. Moreover $x_1 + (x_4 - x_1) + (1 - x_4) = 1$. Therefore $x_1 + \frac{x_1}{\alpha} + x_1 \leq 1$, which implies that $x_1 \leq \frac{\alpha}{2\alpha+1}$. Therefore

$$\begin{aligned} 1 + \frac{x_1(\frac{1}{1+\alpha})}{\frac{1}{2} - \frac{x_1}{2\alpha}} &\leq 1 + \frac{\frac{\alpha}{2\alpha+1}(\frac{1}{1+\alpha})}{\frac{1}{2} - \frac{1}{2(2\alpha+1)}} = 1 + \frac{\frac{\alpha}{2\alpha+1}(\frac{1}{1+\alpha})}{\frac{\alpha}{2\alpha+1}} \\ &= 1 + \frac{1}{1+\alpha} = \frac{2+\alpha}{1+\alpha}. \end{aligned}$$

Now let $\max\{1 - x, x - \frac{x_1}{\alpha}\} = x - \frac{x_1}{\alpha}$. In the calculations we use the fact that $x \geq x_4$ and $x_4 = \frac{1+\alpha}{\alpha}x_1$ (recall that $x_1 = \alpha(x_4 - x_1)$).

$$\begin{aligned} \frac{w(x)}{I(x)} &\leq \frac{x - \frac{x_1}{\alpha(1+\alpha)}}{x - \frac{x_1}{\alpha}} = \frac{x - \frac{x_1}{\alpha} + \frac{x_1}{\alpha} - \frac{x_1}{\alpha(1+\alpha)}}{x - \frac{x_1}{\alpha}} \\ &= 1 + \frac{x_1(\frac{1}{\alpha} - \frac{1}{\alpha(1+\alpha)})}{x - \frac{x_1}{\alpha}} \leq 1 + \frac{x_1(\frac{1}{1+\alpha})}{x_4 - \frac{x_1}{\alpha}} \\ &= 1 + \frac{x_1(\frac{1}{1+\alpha})}{\frac{1+\alpha}{\alpha}x_1 - \frac{1}{\alpha}x_1} = 1 + \frac{1}{1+\alpha} = \frac{2+\alpha}{1+\alpha}. \end{aligned}$$

This proves that for all $x \in (x_4, 1]$ the ratio $\frac{w(x)}{I(x)}$ is at most $\frac{2+\alpha}{1+\alpha}$. □